

Lecture 8

Plan:

- Faces of Polyhedra
- State tons of facts
- Prove them

Faces of Polyhedra

Def: $a^{(1)}, \dots, a^{(k)} \in \mathbb{R}^n$ are

affinely independent if

$$\sum_{i=1}^k \lambda_i a^{(i)} = 0$$

and $\sum \lambda_i = 0$ imply $\lambda_1 = \dots = \lambda_k = 0$.

(w/out $\sum \lambda_i = 0$, is just linear indp.)

linear independence \rightarrow affine independent.

Note: $\text{aff}(X) = \text{lower dim. affine space containing } X$.

$\{c(i)\}$ n. o. independent iff

$\left\{ \alpha \rightarrow \text{affinely independent} \right.$

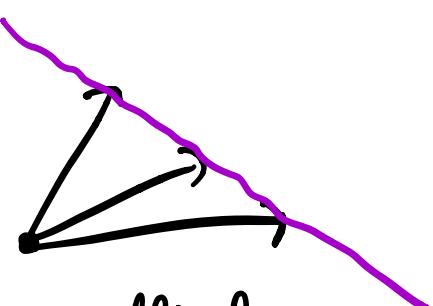
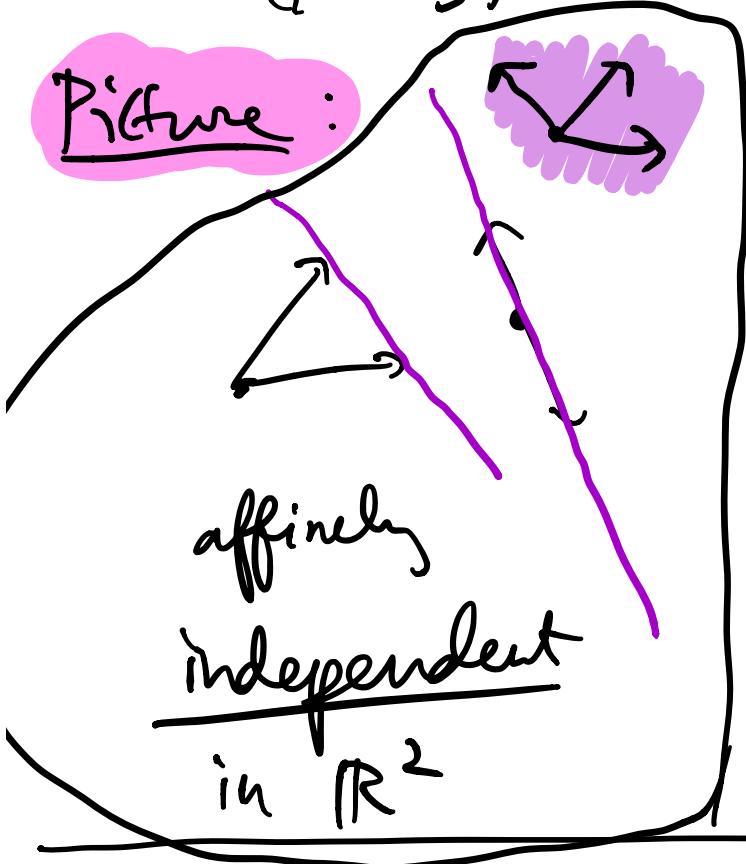
$$\left\{ \begin{bmatrix} \alpha^{(i)} \\ 1 \end{bmatrix} \right\}$$

linearly independent.

$\Leftrightarrow \text{aff}\left(\{\alpha^{(i)}\}\right)$ has dimension $K-1$

vectors.

Picture:



affinely
dependent
in \mathbb{R}^2 .

Def Dimension $\dim(P)$ of

polyhedron P :

$-1 + \max \# \text{affinely}$
 $\text{independent points in } P.$

Equivalently, dimension of
affine hull $\text{aff}(P)$.

Example: $P = \emptyset, \dim(P) = -1$

$P = \text{singleton} \quad \cdot \quad \dim(P) = 0$

$P = \text{line segment} \quad \nearrow \quad \dim(P) = 1$

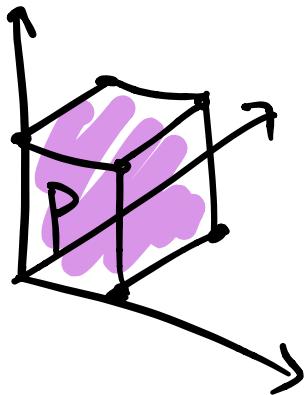
;

$$\text{aff}(P) = \mathbb{R}^n$$

$$\dim(P) = n;$$

P "full dimensional"

e.g. cube in \mathbb{R}^3 : $\{x : 0 \leq x_i \leq 1\}$



$$\dim P = 3$$

$$\dim \mathbb{R}^3 = 3$$

(as polyhedron).

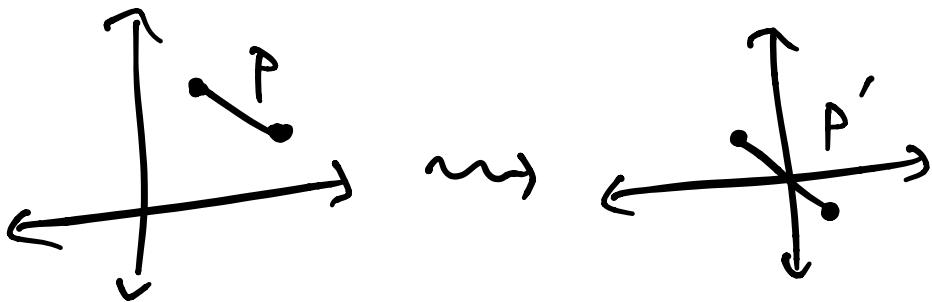
Why affine, not linear? affine

independence is translation
invariant:

if I used max # lin indep points - 1

$$\dim(P) = 1$$

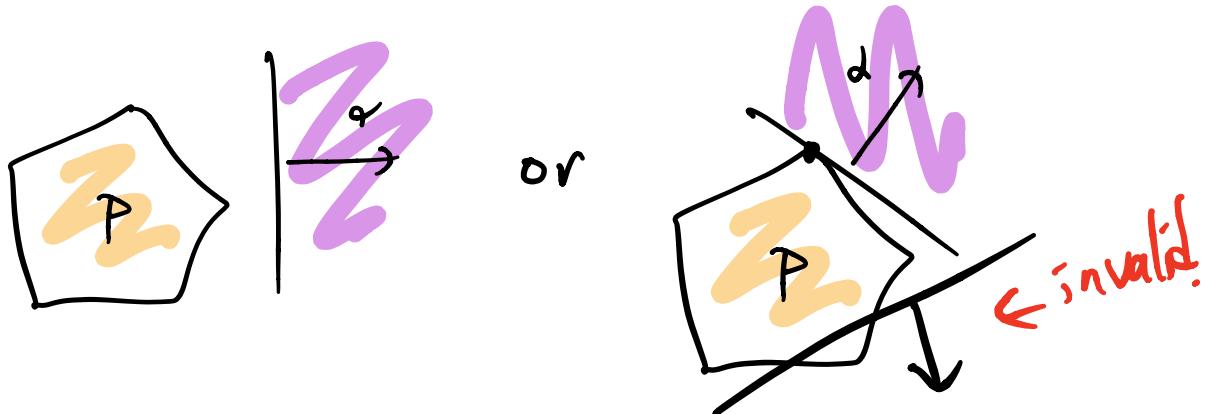
$$\dim(P') = 0$$



$$l = \dim(P) = \dim(P').$$

Def: $\alpha^T x \leq \beta$ is a valid inequality

for P if $\alpha^T x \leq \beta$ for all $x \in P$.

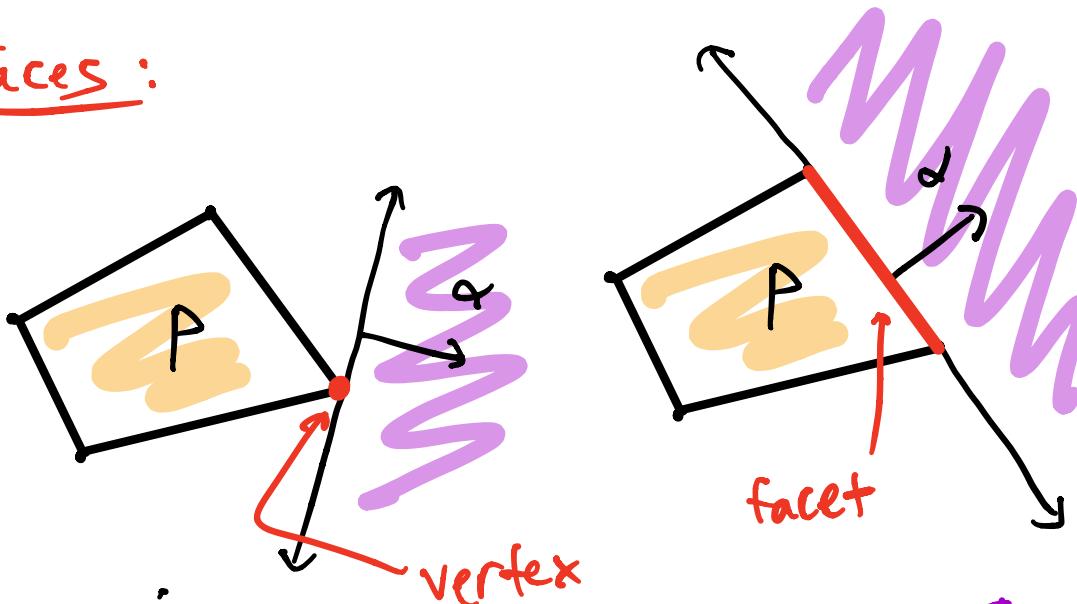


Def A face of a polyhedron

P is $\{x \in P : \alpha^T x = \beta\}$ for

$Q^T x \leq \beta$ valid.

Faces:



Properties:

- Faces are polyhedra
- Empty face & entire P are called trivial faces
- else F nontrivial

$\dim = -1$

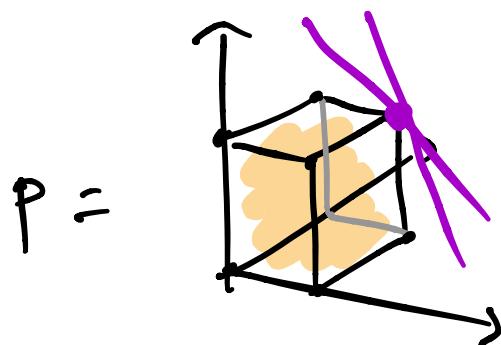
$$0 \leq \dim(F) \leq \dim P - 1$$

- F : $\dim(F) = \dim(P) - 1$ called facets.
- $x \in \text{Int}(F) \cap \text{vert}$ called vertices

• T - array for unitary vector

Ex : list the 28 faces of the cube

$$P = \{x \in \mathbb{R}^3 : 0 \leq x_i \leq 1\}$$



Fact : ∞ many valid ineqs,
but # faces finite!

EVERYTHING ABOUT POLYHEDRA

$$A = \begin{bmatrix} - & a_1^T & - \\ \vdots & \vdots & \vdots \\ - & a_m^T & - \end{bmatrix} \quad P = \{x : Ax \leq b\} \subseteq \mathbb{R}^n$$

Face Characterization:

Any nonempty face of P

is

$$F_I \left\{ x : \begin{array}{l} a_i^T x = b_i \quad \forall i \in I \\ a_i^T x \leq b_i \quad \forall i \notin I \end{array} \right\}$$

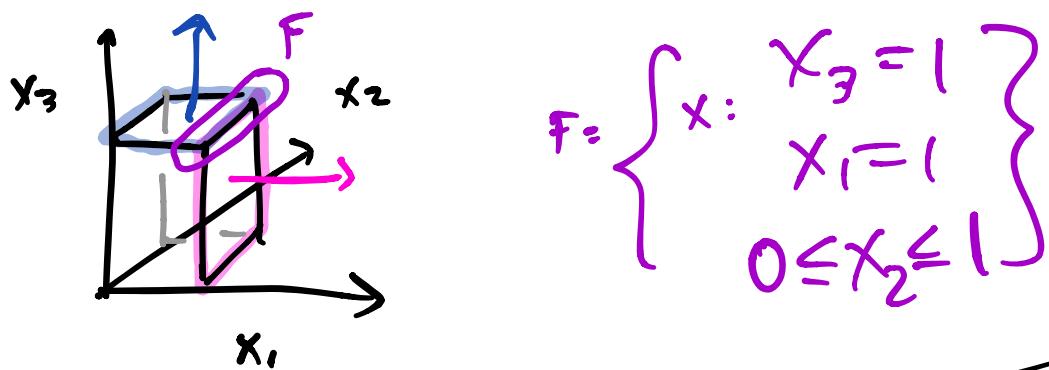
for some set $I \subseteq \{1, \dots, m\}$.

(and F_I is always a face).

\rightsquigarrow intersection of L_{m+1}, \dots, L_n

number of faces \Rightarrow faces

E.g. cube



$$\Rightarrow \# \text{faces} \leq 2^m + 1$$

② Facet Maximality: The facets are the maximal nontrivial faces of a nonempty polyhedron P.

For vertices: just need equalities.

vertices = extreme points. Exercise

③ Vertex Characterization:

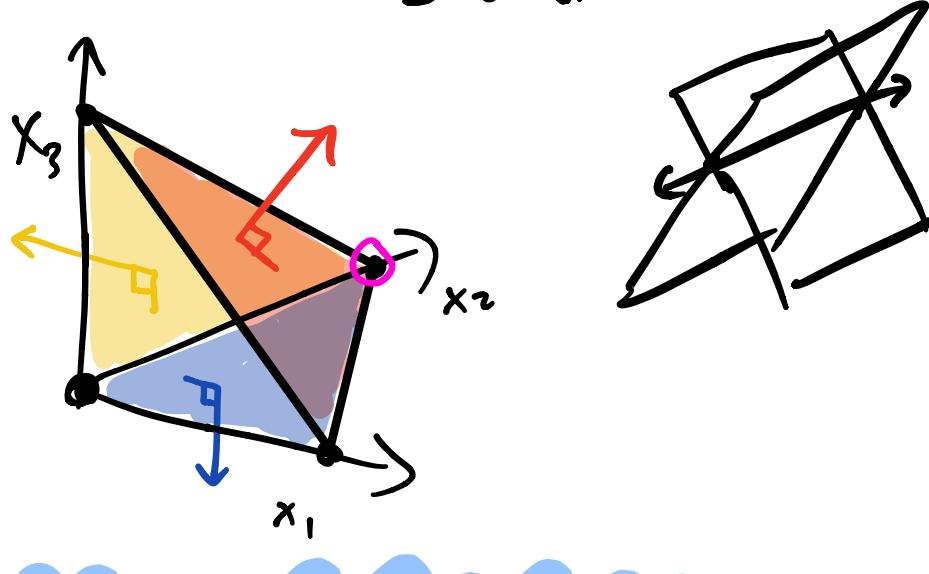
Suppose x^* extreme point of P

Then $\exists I$ s.t. x^* is
the unique soln to

* $a_i^T x = b_i$; for all $i \in I$.

moreover any $x \in P$ that uniquely solves *
is extreme.

e.g. simplex $(0,1,0)$ is intersection of
3 constraints



- Vertex minimality: For $\text{rank}(A) = n$, minimal nontrivial faces of polyhedron P are the vertices.

Exercise: if $\text{rank}(A) < n$, no vertices!

- Polytopes = convex hulls

If a polyhedron P is bounded then $P = \text{conv}(\{\text{extreme points of } P\})$.

(special case of Krein-Milman theorem: compact convex subset of \mathbb{R}^n is $\text{conv}(\text{extreme pts})$).

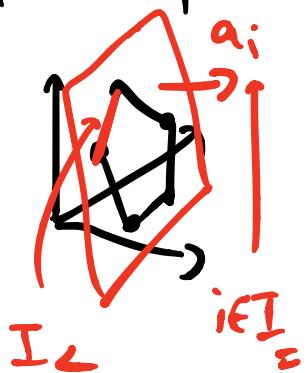
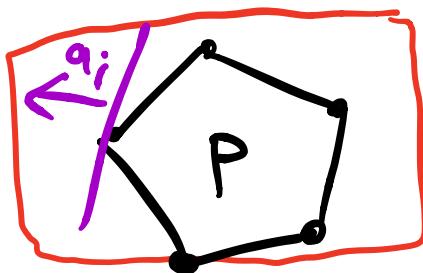
- Facets Characterize



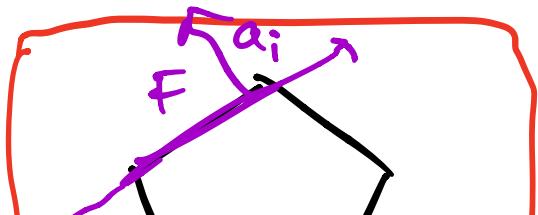
- inequality $a_i^T x \leq b_i$; redundant if P unchanged when it's removed. "equalities"
- $I_0 := \{i : a_i^T x = b_i \text{ } \forall x \in P\}$
- $I_< := \{i : \exists x \in P \text{ } a_i^T x < b_i\}$. "real inequalities"

THEN:

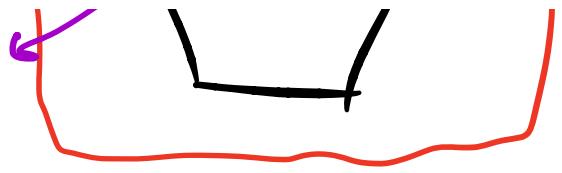
(Sufficiency:) If face $a_i^T x \leq b_i$ for $i \in I_<$ is not facet, then $a_i^T x \leq b_i$ is redundant.



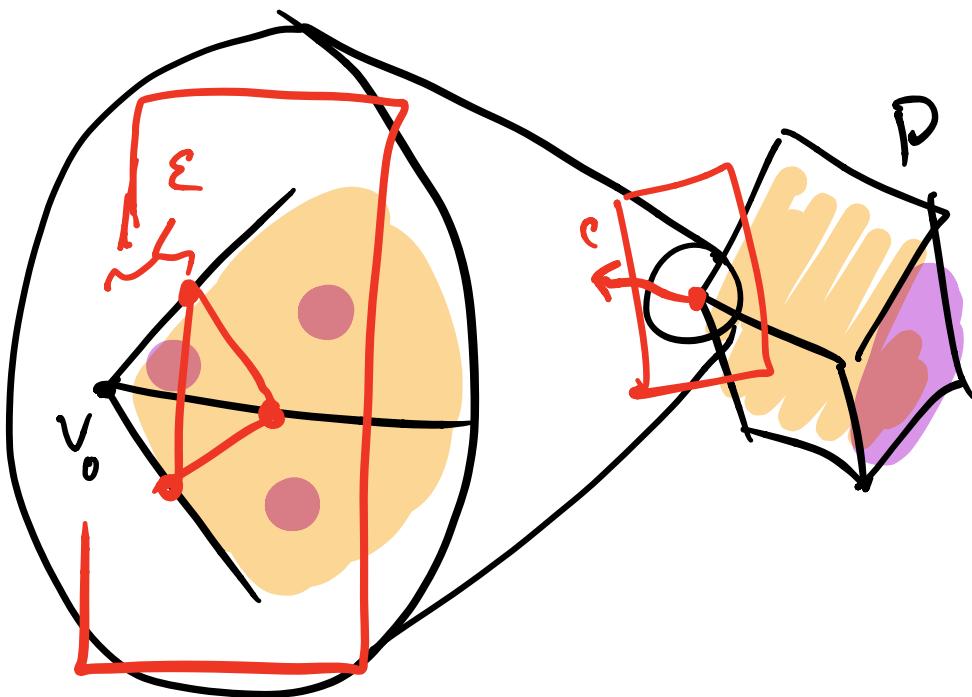
(Necessity:) If F is facet of P, $\exists i \in I_<$ such that F is induced by



$$a_i^T x = b_i$$



• Near vertices =
Cones over polytopes



Let v_0 vertex of P from
valid inequality $c^T x \leq m$.

Let ϵ be such that $c v' \leq m - \epsilon$

for all other vertices v' .

Then

$$P_0 = \{x \in P : c^T x = M - \varepsilon\}$$

is a polytope & is bijection

$$\{P_0's \dim k \text{ faces}\}$$

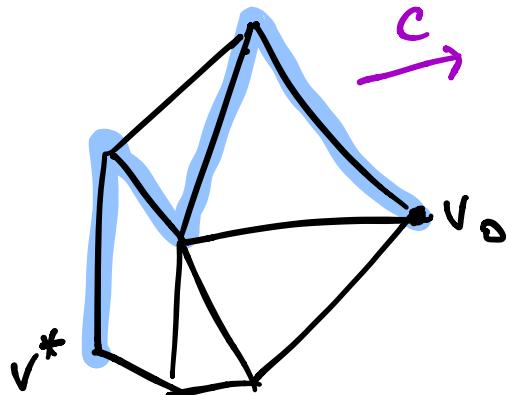


$$\{P's \dim k+1 \text{ faces contain } v_0\}$$

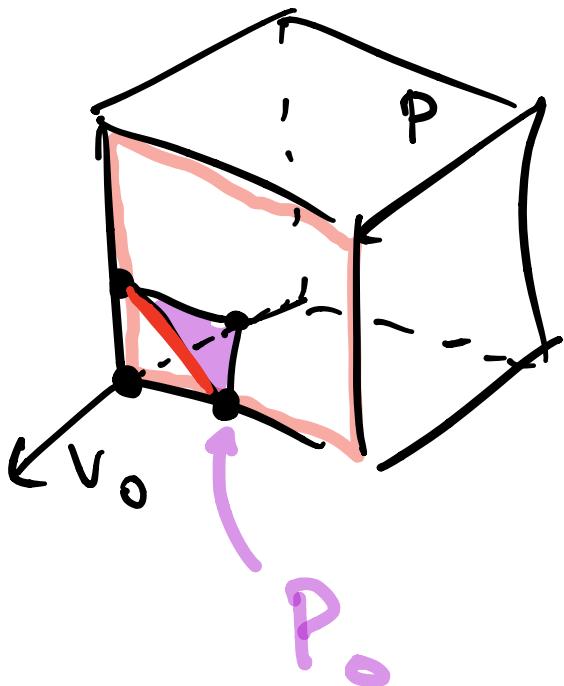
- P's "graph" connected: Graph of vertices & edges of polyhedron P

is always connected.

In particular: if v^* minimum of $c^T x$ over P ,
 v_0 vertex, $\exists v_0 \rightarrow v^*$ path which
decreases objective.



$$v_0 \vec{1} \geq 0$$



bijection: dim k
face F of P_0

dim $k+1$ face of
 P_0 in P containing F
and v_0 .



PROOFS

Recall face characterization:

Let $A \in \mathbb{R}^{m \times n}$, $A = \begin{bmatrix} & \vdots \\ - & a_1^T & - \\ & \vdots \end{bmatrix}$

Any nonempty face of $P = \{x : Ax \leq b\}$

is

$$\left\{ \begin{array}{l} x : a_i^T x = b; \quad \forall i \in I \\ a_i^T x \leq b; \quad \forall i \notin I \end{array} \right\}$$

for some set $I \subseteq \{1, \dots, m\}$.

Proof of converse: exercise

Proof

Consider valid inequality

$$a^T x \leq b$$

giving nonempty face F .

$$F = \{x : a^T x = b\} \cap P$$

- $F = \underline{\text{optimum solutions to bounded LP}}$

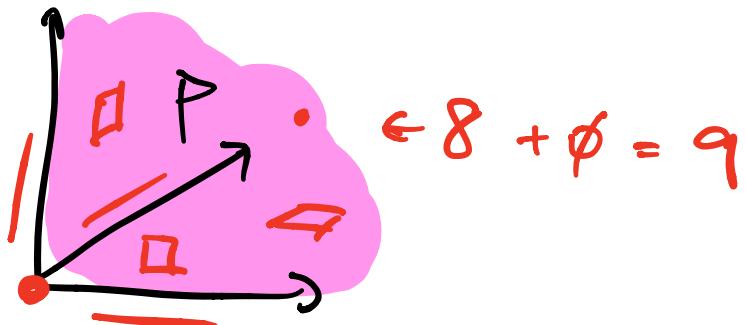
$$\begin{aligned} & \max a^T x \\ (\text{P}) \quad & \text{subject to } Ax \leq b \end{aligned}$$

- Let y^* optimal solution to dual.
 $y^* = (y_1, \dots, y_m)$
- Complementary Slackness:
optimal solns F are

$$\{x : a_i^T x = b_i \text{ for } i : y_i^* > 0\}.$$

Thus we can take $I = \{i : y_i^* > 0\}$. \square

- Ex :
- positive orthant $\{x \in \mathbb{R}^n : x_i \geq 0\} = P$
 - has $2^n + 1$ faces $\uparrow n$ inequalities
 - How many of dim P ? $\{x_i = 0 : i \in I\}$



For polytopes can also bound
faces in terms of # vertices.
(“upper bound theorem”)

“Dehn-Sommerville equations”)

Facet Maximality

Pf : Exercise to prove from face characterization.

Recall vertex characterization:

Let x^* extreme point for P .

Then $\exists I$ s.t. x^* is the unique soln to

$$a_i^T x = b_i \quad \forall i \in I.$$

moreover, any such unique solution $x^* \in P$ is extreme.

Proof: Given extreme point x^* ,

- define $I = \{i : a_i^\top x^* = b_i\}$.

- Note for $i \notin I$, $a_i^\top x^* < b_i$.

- By "faces characterisation", x^* uniquely defined by

$$F = \left\{ \begin{array}{ll} (*) \quad a_i^\top x = b_i & i \in I \\ (***) \quad a_i^\top x \leq b_i & i \notin I \end{array} \right\} = \{x^*\}$$

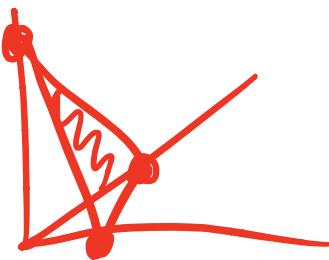
- Suppose \exists other soln. \hat{x} to (*).
(for contradiction)

- Because $a_i^\top \hat{x} \leq b_i$ for $i \notin I$,

$$(1-\varepsilon)x^* + \varepsilon \hat{x}$$

still satisfies $(*)$, $(**)$ for ε small enough.

- Contradicts F having only one point. \square .
-



Basic Feasible Solutions:

For $Q = \{Ax=b, x \geq 0\}$

can describe extreme points
very explicitly.

(every P can be put in this form).

Corollary of Vertex Thm: Extreme pts. of Q as above come from setting

$$x_j = 0 \text{ for } j \in J$$

and finding unique solution to $Ax = b$ for remaining variables.

Can say more: Extreme points

of Q as above are

the basic feasible solutions (bfs),

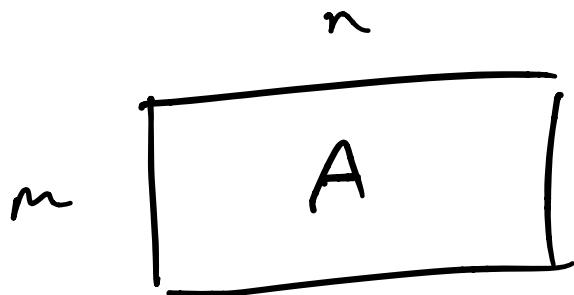
feasible solns obtained as follows:

FILLED IN LEC 7 HANDOUT

- Remove redundant rows

from A ()

)



- Choose m columns of A , C

$$\begin{matrix} & n \\ m & \boxed{} \end{matrix} \quad \parallel \quad = \begin{bmatrix} b \end{bmatrix}$$

- Solve $A_B X_B = 0$,
set

$$x_i^* = \begin{cases} & i \in B \\ & \text{else} \end{cases}$$

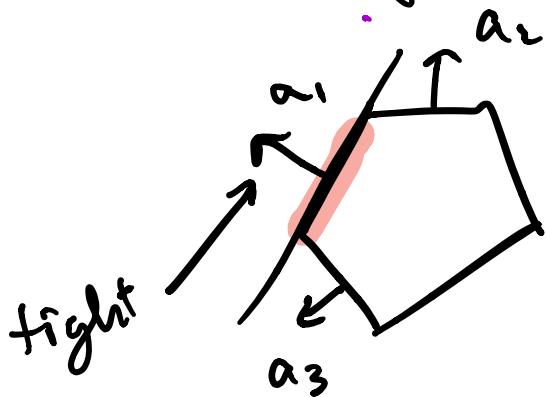
$$\{ \text{bfs} \} = \{ \text{extreme pts} \}.$$

Recall vertex minimality

If $\text{rank } A = n$, vertices are minimal nontrivial facets of P .

$$P = \{x : Ax \leq b\}$$

$$\begin{bmatrix} \vdots \\ -a_i^\top \\ \vdots \end{bmatrix}$$



Proof: Let F min'l face of P .

• Face characterization $\Rightarrow \exists I$

$$\cap_{i \in I} a_i^\top x = 1 \quad (\forall x \in F)$$

$$F = F_I = \left\{ x : \begin{array}{l} a_i^T x = b_i \quad \forall i \in I \\ a_j^T x \leq b_j \quad \forall j \notin I \end{array} \right\}$$

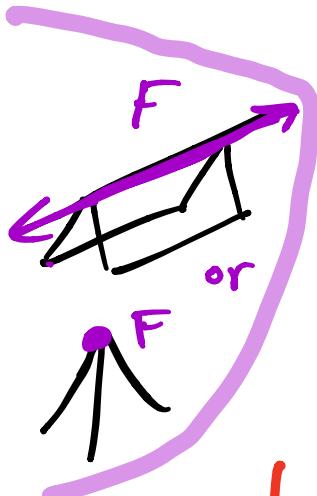
assume no redundant inequalities in I .

and adding any elt to I makes F_I empty. (b/c else F_I face $\subseteq F$).

- Consider two cases:

(a) Only the equalities are needed
 $\underline{\{x : a_i^T x = b_i; \forall i \in I\}}$
 i.e. F is exactly
 $\underline{(a_j^T x \leq b_j \text{ redund. for } F)}$

$$\{x : a_i^T x = b_i; \forall i \in I\}.$$



* Claim: $\forall j \notin I$,

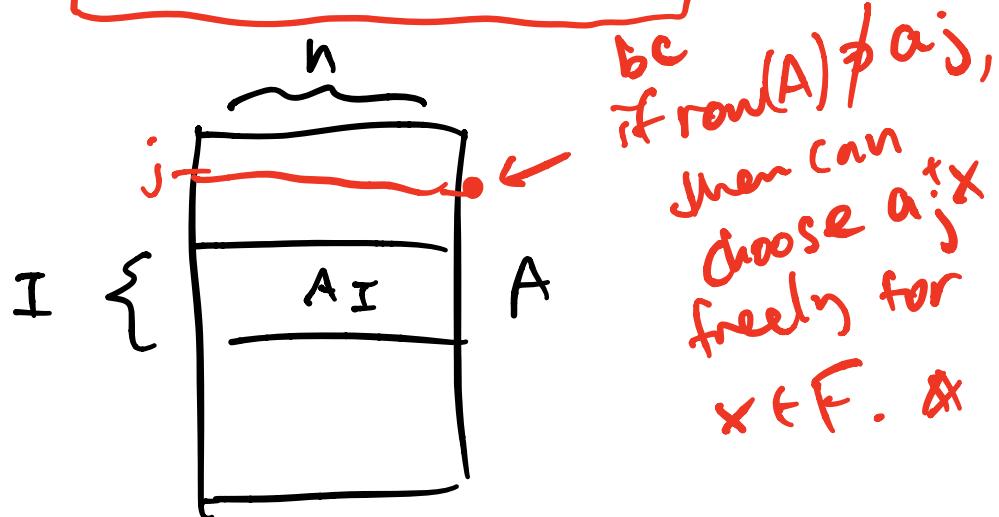
$$a_j \notin \text{lin}(a_i : i \in I).$$

else $a_j^T x = b_j + 1$ has solution
 in F , contradicting $a_j^T x \leq b_j$.

$(a_i^T x : i \in I)$ do not determine
 $a_j^T x$ unless $a_j \in \text{im}(a_i : i \in I)$

* Equivalently: submatrix A_I w/
 rows in I satisfies

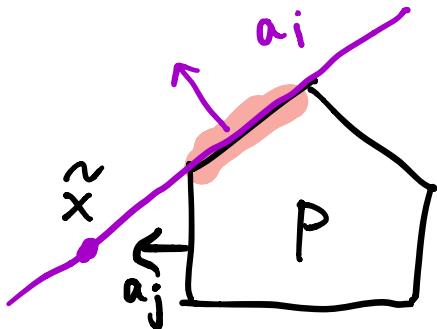
$$\text{row}(A) = \text{row}(A_I) ;$$



hence $\text{rank}(A_I) = \text{rank}(A) = n.$

* Thus: $a_i^T x = b_i$ for $i \in I$
 has unique soln, so F is
 single point, i.e. a vertex. ✓

(b) Some inequality needed:
we'll show is contradiction.



- $\exists j \in I, \hat{x} \text{ w/ } (\hat{x} \notin P)$

$$\begin{aligned} a_i^T \hat{x} &= b_i; i \in I, \\ a_j^T \hat{x} &> b_j \end{aligned}$$

- F nontrivial $\Rightarrow \exists \hat{x} \in F$.

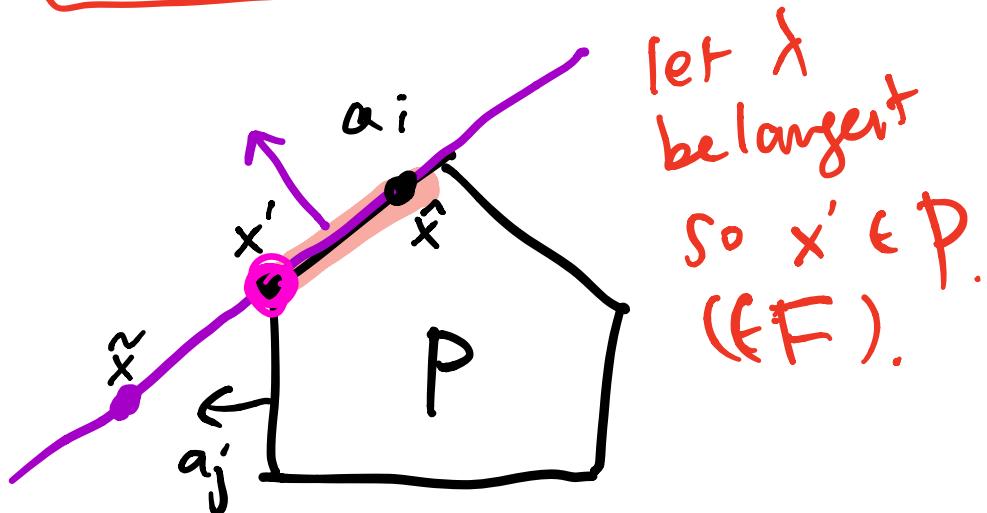
\hat{x} satisfies

$$a_i^T \hat{x} = b_i; i \in I,$$

$$a_j^T \hat{x} \leq b_j$$

- Consider convex combination

$$x' = \lambda \tilde{x} + (1-\lambda) \hat{x}$$



- x' satisfies one more equality

(else could increase λ)
contradicts minimality of F . \square

Finally we can show
equiv b/w bounded polyhedra &
convex hulls. (polytopes).

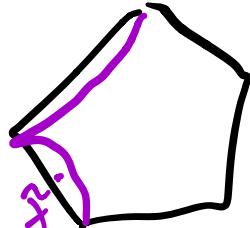
Recall: $P = \{Ax \leq b\}$ bounded

then $P = \text{conv}(X)$ (extreme pts. of P)

i.e. P is polytopal.

Proof: Use TOTIA.

- $X \subseteq P \Rightarrow \text{conv}(X) \subseteq P$.
- Assume for contradiction that $\text{conv}(X) \not\subseteq P$.
- Let $\tilde{x} \in P \setminus \text{conv}(X)$.



• Then

$$\sum_{v \in X} \lambda_v v = \tilde{x}$$

$$\sum_{v \in X} \lambda_v = 1$$

$\lambda_v \geq 0$

has no solution.

• TOTA \Rightarrow

$$\tilde{A} \rightarrow \left[\begin{array}{c|ccccc} \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots \\ \hline \dots & \dots & \dots & \dots & \dots \\ -I & & & & & \end{array} \right] \left[\begin{array}{c|c} I & \\ \hline \lambda & \end{array} \right] = \left[\begin{array}{c|c} x_1 & \\ \hline x_2 & \\ \hline 1 & \\ \hline 0 & \end{array} \right]$$

\leq

$\Delta \rightsquigarrow \square$
 $\Leftarrow \Rightarrow \geq$
 $\equiv \Rightarrow ?$

has no soln \Leftrightarrow

$$\tilde{A}^T y = 0, \tilde{b}^T y < 0, y \neq 0$$

has soln. i.e.

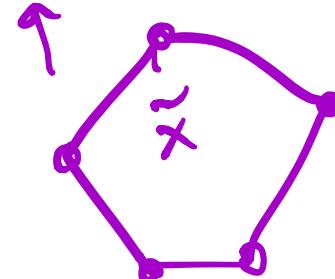
$$\tilde{A}^T \quad y$$

,

$$\begin{pmatrix} \vdots & \vdots \\ -v^T s & | & \vdots \\ \vdots & \vdots & \vdots \\ -I \end{pmatrix}$$

$$\begin{matrix} c \\ t \\ s \end{matrix}$$

$\Rightarrow (*)$



$$y^T \vec{0}$$

for $s \geq 0$, and $y^T b \leq 0$.

(**)

I.E.

$$\hookrightarrow (*) \quad t + c \cdot v \geq 0 \quad \forall v \in X$$

$$(**) \quad t + c \cdot x \leq 0. \quad \Rightarrow c \cdot x < c \cdot v$$

- P bounded \Rightarrow

$$\min \{c^T x : x \in P\} = z^* > -\infty.$$

- Face induced by nonempty, but contains no vertex.

(because \nRightarrow objective is less on \tilde{x} than any vertex.)

- Contradicts vertex minimality! ✓

~~C~~ applies b/c $\text{rank } A = n$;

if $\text{rank } A < n$, P
not bounded (b/c some)

solution to $Ay = 0$. \blacksquare .

assume wlog $0 \in P \Rightarrow b_i \geq 0$

